[Project Writeup] Traffic Sign Classification 09-Sep-2017

Files Submitted

| CRITERIA | MEETS SPECIFICATION |
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| Submission Files | The Submission includes |
| | Traffic_Sign_Classifier.ipynb files testSet/ test images Writeup file as writeup.pdf |

Dataset Exploration

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| Dataset Summary | Database includes 3 files |
| | Train.p - training pickle file valid.p - validation pickle file Test.p - test pickle file |
| | Each file is a pickle file, It has features and labels in keys `features` and `labels` |
| Exploratory Visualization | The submission includes an exploratory visualization on the dataset. |
| | Training data has 43 traffic signs, each sign differs in the number of samples available for training. |
| | l: 41 c: 210 |
| | l: 29 c: 240 |
| | l: 7 c: 1290 |
| | l: 28 c: 480 |
| | |

Design and Test a Model Architecture

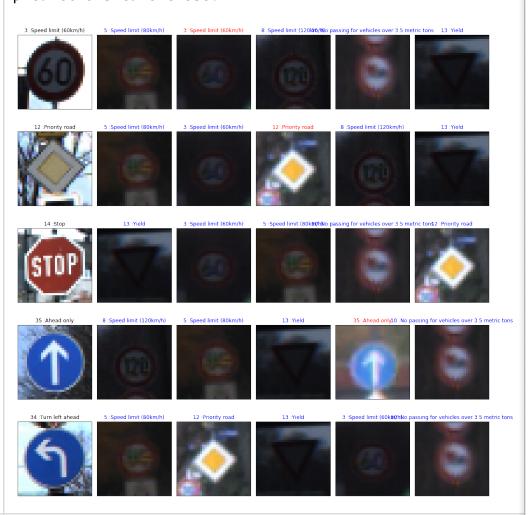
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| Preprocessing | • No pre processing where used. |
| Model Architecture | The submission provides details of the characteristics and qualities of the architecture, including the type of model used, the number of layers, and the size of each layer. Visualizations emphasizing particular qualities of the architecture are encouraged. |
| Model Training | Learning rate of 0.001 was used. LeNet architecture was used for training a data set with images of shape (32, 32, 3) EPOCHS = 10 AdamOptimizer a variant of Stochastic Gradient Descent was used as optimiser. |
| Solution Approach (The submission describes the approach to finding a solution. Accuracy on the validation set is 0.93 or greater.) | Training Accuracy = 0.987 Validation Accuracy = 0.885 Test Accuracy = 0.893 |

Test a Model on New Images

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| Acquiring New Images | The submission includes 2 * five new German Traffic signs found on the web. |
| | One set has clear images of the traffic signs. The other set has images taken in dark normal road conditions. |
| Performance on New Images | <pre>Images acquired from: http://benchmark.ini.rub.de/? section=gtsrb&subsection=news</pre> |
| | On a set of sample computer images acquired 20 % accuracy. |
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While the set with road conditions gave 0 accuracy. Many predictions came close.



Model Certainty - Softmax Probabilities The top five softmax probabilities of the predictions on the captured images are outputted. The submission discusses how certain or uncertain the model is of its predictions.

Signs [0, 1, 2, 3, 4]
Sign idx : 0 kresults : [8 5 3 4 6]
Sign idx : 1 kresults : [3 5 6 8 10]
Sign idx : 2 kresults : [5 8 3 10 2]
Sign idx : 3 kresults : [3 5 10 13 8]
Sign idx : 4 kresults : [8 5 4 18 3]

| Signs [3, 12, 14, 35, 34] Sign idx: 3 kresults: [5 3 8 10 13] Sign idx: 12 kresults: [5 3 12 8 13] Sign idx: 14 kresults: [13 3 5 10 12] Sign idx: 35 kresults: [8 5 13 35 10] Sign idx: 34 kresults: [5 12 13 3 10] |
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Suggestions to Make Your Project Stand Out!

Here are a few ideas for going beyond the requirements outlined in the rubric.

AUGMENT THE TRAINING DATA

Augmenting the training set might help improve model performance. Common data augmentation techniques include rotation, translation, zoom, flips, and/or color perturbation. These techniques can be used individually or combined.

ANALYZE NEW IMAGE PERFORMANCE IN MORE DETAIL

Calculating the accuracy on these five German traffic sign images found on the web might not give a comprehensive overview of how well the model is performing. Consider ways to do a more detailed analysis of model performance by looking at predictions in more detail. For example, calculate the <u>precision and recall</u> for each traffic sign type from the test set and then compare performance on these five new images..

If one of the new images is a stop sign but was predicted to be a bumpy road sign, then we might expect a low recall for stop signs. In other words, the model has trouble predicting on stop signs. If one of the new images is a 100 km/h sign but was predicted to be a stop sign, we might expect precision to be low for stop signs. In other words, if the model says something is a stop sign, we're not very sure that it really is a stop sign.

Looking at performance of individual sign types can help guide how to better augment the data set or how to fine tune the model.

CREATE VISUALIZATIONS OF THE SOFTMAX PROBABILITIES

For each of the five new images, create a graphic visualization of the soft-max probabilities. Bar charts might work well.

VISUALIZE LAYERS OF THE NEURAL NETWORK

See Step 4 of the Iptyon notebook for details about how to do this.